

ON A MODEL FOR COMPUTING ROUND-OFF ERROR OF A SUM

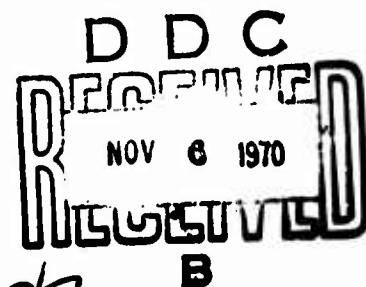
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Given real numbers a_1, a_2, \dots, a_n , we are interested in the classic problem of the error in computing $S = \sum_{i=1}^n a_i$ when the sum is computed by $\tilde{S}_0 = \sum_{i=1}^n a_i^*$ where a_i^* is the nearest integer to a_i . We shall first study this error as a function of a Δ shift, i.e., when all numbers a_i are each shifted Δ and then rounded;

$$(1) \quad S - n\Delta = \sum_{i=1}^n (a_i - \Delta)$$

$$(2) \quad \tilde{S}_\Delta - n\Delta = \sum_{i=1}^n (a_i - \Delta)^*$$

We will then let Δ become a random variable that can take on uniformly any value in the interval $-\frac{1}{2} \leq \Delta \leq +\frac{1}{2}$. Different choices of Δ give rise to different rounding errors $\tilde{S}_\Delta - S$ and the variance of the distribution of $\tilde{S}_\Delta - S$ can be used to measure the variability of the rounding error due to the random selection of the origin of the real numbers a_i with respect to that of the computer.

The cumulative error from (1) and (2) is

$$(3) \quad \tilde{S}_\Delta - S = \sum_{i=1}^n [(a_i - \Delta)^* - (a_i - \Delta)]$$

Let f_i be the positive fractional part of a_i and let a_i^* be the largest integer not exceeding a_i , i.e.,

$$(4) \quad a_i = a_i^* + f_i$$

Denoting by r_i the error of the i^{th} term, we have

$$(5) \quad r_i = [(a_i - \Delta)^* - (a_i - \Delta)] = \begin{cases} 1 - (f_i - \Delta) & \text{if } -\frac{1}{2} \leq \Delta \leq -\frac{1}{2} + f_i \\ -(f_i - \Delta) & \text{if } -\frac{1}{2} + f_i \leq \Delta \leq +\frac{1}{2} \end{cases}$$

To prove the above, we note that $f_i - \Delta = (a_i - \Delta) + e_i$. If $-\frac{1}{2} \leq f_i - \Delta \leq +\frac{1}{2}$ then $(a_i - \Delta)$ is rounded to a_i . Hence $a_i - \Delta$ is rounded down if $-\frac{1}{2} + f_i \leq \Delta$ otherwise rounded up.

Denoting expected value by E , we have by direct evaluation

$$(6) \quad E(r_i) = \int_{-\frac{1}{2}}^{+\frac{1}{2}} r_i d\Delta = 0$$

Assume $f_i \leq f_j$, then

$$\begin{aligned} E(r_i r_j) &= \int_{-\frac{1}{2}}^{-\frac{1}{2}+f_i} r_i r_j d\Delta + \int_{-\frac{1}{2}+f_i}^{-\frac{1}{2}+f_j} r_i r_j d\Delta + \int_{-\frac{1}{2}+f_j}^{+\frac{1}{2}} r_i r_j d\Delta \\ &= \int_{-\frac{1}{2}}^{+\frac{1}{2}} (f_i f_j - \Delta(f_i + f_j) + \Delta^2) d\Delta \\ &\quad + \int_{-\frac{1}{2}}^{-\frac{1}{2}+f_i} [(1-f_i-f_j) + 2\Delta] d\Delta \\ &\quad + \int_{-\frac{1}{2}+f_i}^{-\frac{1}{2}+f_j} [-f_i + \Delta] d\Delta \end{aligned}$$

Performing indicated integration yields:

$$(7) \quad E(r_i r_j) = \frac{1}{2} [|f_j - f_i|^2 - |f_j - f_i| + \frac{1}{6}]$$

which is one-half the 2nd order Bernoulli Polynomial in $|f_j - f_i|$. For $f_j < f_i$ we also get (7). Note that the individual errors r_i and r_j are not independent of one another.

It now follows that

$$(9) \quad E(\tilde{S}) = S$$

$$(10) \quad E(\tilde{S}-S)^2 = E\left(\sum_{i=1}^n \sum_{j=1}^n r_i r_j\right) = \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n [|f_i - f_j|^2 - |f_i - f_j| + \frac{1}{6}]$$

The usual value of variance, $E(\tilde{S}-S)^2 = n/12$, will result if we further assume f_i are independently drawn from uniform distributions on $[0 \leq f_i \leq 1]$.

Theorem: If the fractional parts of all a_i are equal to each other, then each term of (10) is maximum for $0 \leq f_i \leq 1$ and

$$(11) \quad \text{Max } E(\tilde{S}-S)^2 = \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \left(\frac{1}{6}\right) = \frac{n^2}{12} .$$

From (10) we have an interesting inequality, namely for all f_i

$$(12) \quad V(f) = \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n \{|f_i - f_j|^2 - |f_i - f_j| + \frac{1}{6}\} \geq 0$$

This function is not convex even for $n=2$, since $f^{(1)} = (\frac{1}{2}, 0)$ and $f^{(2)} = (-\frac{1}{2}, 0)$ yields $V(f^0) = V(f^1) = \frac{1}{12} + \frac{1}{12} - \frac{1}{12} = \frac{1}{12}$ but

$V\left(\frac{f^1 + f^2}{2}\right) = V(0) = \frac{3}{12}$. There appears to be no obvious direct way to establish that $V(f) \geq 0$ for all $0 \leq f_i \leq 1$. Our development shows $V(f)$ to be a variance and this, of course, constitutes an indirect proof. We can replace (12) by a convex realization: Assume $f_i \geq f_{i+1}$ for all i , then the problem of finding $\text{Min } V(f)$ can be rewritten:

$$(13) \quad \text{Find Min } [V(f)] = \sum_{i < j} (f_i - f_j)^2 + \frac{n^2}{12} - [(n-1)f_1 + (n-3)f_2 + (n-5)f_3 + \dots + (n-2k+1)f_k + \dots - (n-1)f_n]$$

subject to

$$(14) \quad f_1 \geq f_2 \dots \geq f_n$$

$$(15) \quad 0 \leq f_i \leq 1$$

Formally (13), (14), (15), is a positive definite quadratic program. Fortunately, as we shall see this can be solved by classical calculus by ignoring inequalities (14) and (15).

Theorem: Equally spaced $f_i = (n - i)/n$, ($i = 1, \dots, n$) yields $\text{Min } V(f) = \frac{1}{12}$ independent of n , i.e., the variance of the sum in this case is minimum and is the same as the variance of the individual terms forming the sum.

Proof: Setting partials = 0 in (13) yields:

$$(16) \left\{ \begin{array}{cccccc} 2(n-1)f_1 & -2f_2 & \dots & -2f_n & = & (n-1) \\ -2f_1 & +2(n-1)f_2 & & -2f_n & = & (n-3) \\ -2f_1 & -2f_2 \dots 2(n-1)f_{n-1} & -2f_n & = & -(n-3) \\ -2f_1 & -2f_2 & & 2(n-1)f_n & = & -(n-1) \end{array} \right.$$

Adding shows the equations to be dependent. Hence we may drop the last equation as redundant. Moreover, we can always translate the f_i so that the smallest f_i , namely $f_n = 0$

Re-adding yields:

$$2f_1 + 2f_2 + \dots + 2f_{n-1} + 0 = (n-1), f_n = 0.$$

Adding this last equation to each of the others gives

$$2nf_i = (n - 2i + 1) + (n - 1) = 2(n - i)$$

$$(17) f_i = (n - i)/n$$

Evidently the conditions $0 \leq f_i \leq 1$ and $f_i \geq f_{i+1}$ are (by good luck) also satisfied so that (17) yields the minimum, namely

$$(18) \text{Min } V(f) = \frac{n^2}{12} - \frac{1}{2} \sum_{i=1}^n (n-2i+1)f_i = \frac{1}{12} .$$